Uplink Interference Management in Massive MIMO Enabled Heterogeneous Cellular Networks
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Uplink Interference Management in Massive MIMO Heterogeneous Cellular Networks

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Abstract—Considering the fact that massive multiple input multiple output (MIMO) macrocells and large number of small cells are deployed in the heterogeneous cellular networks for 5G, more users will be supported for uplink transmissions, which in turn brings more severe uplink interference. As a result, interference management is crucial for enhancing the uplink performance in these networks. In this letter, we exploit uplink power control and cell range expansion for interference coordination. Our results show that the uplink performance in the massive MIMO macrocells can be significantly improved through uplink power control in the small cells, while more uplink transmissions in the macrocells have mild adverse effect on the uplink performance of the small cells. In addition, the uplink performance of the massive MIMO macrocells can be improved by expanding the small cell range when more users are served in the macrocells.

Index Terms—Massive MIMO, heterogeneous cellular networks, uplink interference management.

I. INTRODUCTION

Massive multiple input multiple output (MIMO) and heterogeneous cellular network (HCN) are two promising solutions for supporting high data traffic in 5G networks. Massive MIMO, as a large-scale multiuser MIMO transmission strategy, has the capability of communicating with dozens of users at the same time and frequency band [1]. In HCNs, small cell base stations (BSs) are deployed on a large-scale, which allows more users to be served. As such, massive MIMO aided HCNs has recently attracted much attention [2, 3].

In the massive MIMO aided HCNs, downlink transmission has been examined by considering different performance metrics such as spectrum efficiency [2] or energy efficiency [3]. However, interference management is critically important for the uplink transmission, since more users served by the massive MIMO macrocells and large numbers of small cells results in more severe uplink interference. The work of [4] considered the uplink power control for interference mitigation in the two-tier femtocell networks, and two schemes were utilized to suppress the cross-tier interference under a fixed threshold and an adaptive threshold. In [5], Perron-Frobenius theory was applied to obtain the necessary and sufficient conditions for the uplink power control under quality constraint in the massive MIMO systems. In the LTE Release-10, enhanced inter-cell interference coordination (eICIC) was introduced by using the cell range expansion (CRE) to cope with the uplink/downlink imbalance [6].

While the aforementioned literature laid a good foundation in understanding the uplink power control or CRE for coordinating interference in the HCNs or massive systems, interference management for the uplink massive MIMO aided HCNs has not been conducted yet. As such, we propose a joint uplink power control and CRE scheme for uplink interference coordination in a two-tier massive MIMO aided HCNs consisting of picocells and massive MIMO macrocells. Specifically, since users in the massive MIMO macrocells are provided with large array gains, the level of their transmit power can be greatly lowered. Hence, we adopt the uplink power control in the picocells. Due to the massive MIMO array gains, the uplink/downlink imbalance is boosted in the massive MIMO aided HCNs. In light of this, CRE is employed to tackle the uplink/downlink imbalance problem. We characterize the impacts of massive MIMO and HCNs in terms of spectrum efficiency.

II. SYSTEM DESCRIPTION

We consider the uplink transmission in a two-tier heterogeneous networks (HetNets), in which massive MIMO macrocells are overlaid with picocells. The locations of macrocell base stations (MBSs) are modeled following a homogeneous Poisson point process (PPP) $\Phi_M$ with density $\lambda_M$. The locations of picocell base stations (PBSs) are modeled following an independent homogeneous PPP $\Phi_P$ with density $\lambda_P$. Each macrocell BS is equipped with $N$ antennas and receives data streams from $S$ single-antenna users over the same time and frequency band [7, 8], while each PBS equipped with one single antenna receives one data stream from a single-antenna user in each transmission. It is assumed that the density of users is much greater than that of BSs so that there always will be one active mobile user at each time slot in every small cell and multiple active mobile users in every macrocell. The linear zero-forcing beamforming (ZFBF) is employed to cancel the intra-cell interference at the MBS. All the channels undergo independent and identically distributed (i.i.d.) quasi-static Rayleigh fading.

A. Uplink Power Control

Owing to the benefits of massive MIMO such as large antenna array gain, users associated with the MBSs can use lower transmit power, hence we focus on the power control in the picocells. Specifically, the open-loop uplink power control is applied in the picocells, and the transmit power for a user associated with the PBS is given by

$$P_{u,P} = \min \left\{ P_{\text{max}}, P_o L (|X_k|)^{-\eta} \right\},$$

(1)

where $P_{\text{max}}$ is the maximum transmit power, $P_o$ is the normalized power density, $L (|X_k|) = \beta |X_k|^{-\alpha}$ is the path loss with the exponent $\alpha$, and $\beta$ is the frequency dependent constant value, $|X_k|$ is the distance between the user and its associated PBS, $\eta \in [0, 1]$ is the path loss compensation factor, which controls the picocell user’s transmit power. Here $\eta = 1$.
represents that the path loss between a user and its serving PBS is fully compensated, and \( \eta = 0 \) represents that there is no path loss compensation.

### B. User Association

We consider the downlink-centric user association as utilized in the current 4G systems [9]. In the downlink, each MBS transmits \( S \) user-streams with equal power assignment. As such, the long-term average receive power at a macrocell user (MUE) which is connected with the MBS \( \ell \in \Phi_M \) is given by

\[
P_{r,\ell} = \frac{G_a P_M}{S} L(|X_\ell|),
\]

where \( G_a \) is the array gain, \( P_M \) is the MBS’s transmit power, \( L(|X_\ell|) = \beta |X_\ell|^{-\alpha_M} \) is the path loss function, \( |X_\ell| \) is the distance, and \( \alpha_M \) is the path loss exponent. The array gain \( G_a \) obtained by the ZFBF transmission is \( N = S + 1 \) [2, 8]. In the picocell, considering the effect of picocell range expansion, the long-term average receive power at a picocell user (PUE) which is connected with the PBS \( \kappa \in \Phi_P \) is expressed as

\[
P_{r,\kappa} = P_P L(|X_\kappa|) B,
\]

where \( P_P \) is the PBS’s transmit power and \( L(|X_\kappa|) = \beta |X_\kappa|^{-\alpha_P} \) is the path loss function, and \( B \) is the biasing factor, which is useful for offloading the data traffic to small cells in conventional HetNets [10].

### C. Uplink Channel Model

We assume that a typical serving base station is located at the origin \( o \). The receive signal-to-interference-plus-noise ratio (SINR) of a typical serving MBS at a random distance \( |X_{o,M}| \) from its intended MUE is given by

\[
\text{SINR}_M = \frac{P_{u,M} h_{u,M} L(|X_{o,M}|)}{I_M + I_P + \delta^2},
\]

where

\[
I_M = \sum_{i \in \Phi_{u,M} \setminus B(o)} P_{u,M} h_{u,M} L(|X_{i,M}|)
\]

is the uplink intra-tier interference, \( I_P = \sum_{j \in \Phi_{u,P}} P_{u,P} h_{j,M} L(|X_{j,M}|) \) is the uplink inter-tier interference, \( P_{u,M} \) and \( P_{u,P} \) are the transmit power of the MUEs and PUEs, respectively, \( h_{u,M} \sim \Gamma(N - S + 1, 1) \) [8] is the small-scale fading channel power gain between the typical serving MBS and its intended user, \( h_{j,M} \sim \exp(1) \) and \( |X_{j,M}| \) are the small-scale fading interfering channel power gain and distance between the typical serving MBS and interfering MUE \( i \in \Phi_{u,M} \setminus B(o) \) (\( \Phi_{u,P} \setminus B(o) \) is the point process corresponding to the interfering MUEs except the intra-cell users), respectively, \( h_{j,M} \sim \exp(1) \) and \( |X_{j,M}| \) are the small-scale fading interfering channel power gain and distance between the typical serving MBS and interfering PUE \( j \in \Phi_{u,P} \) (\( \Phi_{u,P} \) is the point process corresponding to the interfering PUEs), respectively, and \( \delta^2 \) is the noise power.

The SINR of a typical serving PBS at a random distance \( |X_{o,P}| \) from its intended user is given by

\[
\text{SINR}_P = \frac{P_{o,P} g_{o,P} L(|X_{o,P}|)}{J_M + J_P + \delta^2},
\]

where

\[
J_M = \sum_{i \in \Phi_{o,M}} P_{o,M} g_{i,P} L(|X_{i,P}|), \quad J_P = \sum_{j \in \Phi_{o,P} \setminus B(o)} P_{o,P} g_{j,P} L(|X_{j,P}|), \quad g_{o,P} \sim \exp(1)
\]

is the small-scale fading channel power gain between the typical serving PBS and its intended user, \( g_{i,P} \sim \exp(1) \) and \( |X_{i,P}| \) are the small-scale fading interfering channel power gain and distance between the typical serving PBS and interfering user \( i \in \Phi_{o,M}, \) respectively, \( g_{j,P} \sim \exp(1) \) and \( |X_{j,P}| \) are the small-scale fading interfering channel power gain and distance between the typical serving MBS and interfering user \( j \in \Phi_{o,P} \setminus B(o), \) respectively.

### III. Area Uplink Spectrum Efficiency

In this section, we evaluate the area uplink spectrum efficiency for massive MIMO macrocell tier and picocell tier. The area uplink spectrum efficiency allows us to examine the impacts of different densities of picocells and massive MIMO macrocells on the uplink spectrum efficiency of different tiers.

Based on [10], we first obtain the probability density functions of the distance \( |X_{o,M}| \) between a typical MUE and its serving MBS \( B_{o,M} \) and the distance \( |X_{o,P}| \) between a typical PUE and its serving PBS \( B_{o,P} \) as (6) and (7) at the top of the next page, respectively. Based on the distance statistic properties, we have the following two theorems.

**Theorem 1:** The area uplink spectrum efficiency for the macrocell tier is lower bounded as

\[
R_M = S \lambda_M \log_2 \left( 1 + e^{\Delta_1 + \Delta_2} \right),
\]

where

\[
\Delta_1 = \ln \left( \frac{P_{u,M} \beta}{\psi(N - S + 1)} \right) - \alpha_M \int_0^\infty \ln \left( f_{|X_{o,M}|}(x) \right) dx,
\]

and

\[
\Delta_2 = -\ln \left( \int_0^\infty \mathbb{E}_1(x) f_{|X_{o,M}|}(x) dx + \int_0^\infty \mathbb{E}_2(x) f_{|X_{o,M}|}(x) dx + \delta^2 \right),
\]

where \( \psi(\cdot) \) is the digamma function [11], \( f_{|X_{o,M}|}(x) \) is given by (6), \( \mathbb{E}_1(x) \) and \( \mathbb{E}_2(x) \) are given by (10) and (11) in the following page, \( R_M(x) = \left( \frac{P_{o,M} BS}{\lambda_M (N - S + 1)} \right)^{1/\alpha_P} x^{\alpha_M/\alpha_P} \) and \( r_\alpha = \left( \frac{P_{o,M} BS}{\lambda_M} \right)^{1/(\alpha_P \eta)} \beta^{1/\alpha_P} \).

**Proof:** The proof is provided in Appendix A.

**Theorem 2:** The area uplink spectrum efficiency for the picocell tier is given by

\[
R_P = \frac{\lambda_P}{\ln 2} \int_0^\infty \frac{F_{\text{cov}}(\gamma)}{1 + \gamma} d\gamma,
\]

where

\[
F_{\text{cov}}(\gamma) = \int_0^\infty F_{|X_{o,P}|}(x) f_{|X_{o,P}|}(x) dx
\]

is the complementary distribution function (CCDF) of \( \text{SINR}_P \), here \( f_{|X_{o,P}|}(x) \) is given by (7) and \( F_{|X_{o,P}|}(\gamma) \) is given by (12), in which \( P_{o,P} = \min \{ P_{\text{max}}, P_o (\beta x^{\alpha_P} - \eta) \} \), \( R_P(x) = \left( \frac{P_{o,P} BS}{\lambda_P} \right)^{1/\alpha_P} x^{\alpha_P/\alpha_P} \), \( 2 F_1 \ldots \ldots \cdot F_1 \) is the Gauss hypergeometric function [12, (9.142)], and \( F_1(A) \) is the indicator function that returns one if the condition \( A \) is satisfied. The proof of (12) can be easily provided by following the similar approach in [10].


$$f_{|X_{o,M}|}(x) = \frac{2\pi\lambda_M}{A_M} x \exp \left\{ -\pi \lambda_M x^2 - \pi \lambda_P \left( \frac{P_P BS x^{\alpha_M}}{P_M (N - S + 1)} \right)^{2/\alpha_P} \right\}$$  \hspace{1cm} (6)

$$f_{|X_{o,P}|}(x) = \frac{2\pi \lambda_P}{A_P} x \exp \left\{ -\pi \lambda_P x^2 - \pi \lambda_M \left( \frac{P_M (N - S + 1) x^{\alpha_P}}{P_P BS} \right)^{2/\alpha_M} \right\}$$  \hspace{1cm} (7)

where

$$A_M = 2\pi \lambda_M \int_0^\infty r \exp \left\{ -\pi \lambda_M r^2 - \pi \lambda_P \left( \frac{P_P BS r^{\alpha_M}}{P_M (N - S + 1)} \right)^{2/\alpha_P} \right\} dr$$  \hspace{1cm} (8)

$$A_P = 2\pi \lambda_P \int_0^\infty r \exp \left\{ -\pi \lambda_P r^2 - \pi \lambda_M \left( \frac{P_M (N - S + 1) r^{\alpha_P}}{P_P BS} \right)^{2/\alpha_M} \right\} dr$$  \hspace{1cm} (9)

$$\Xi_1 (x) = \frac{P_{u,M} 2\pi^2 S \lambda_M}{\alpha_M - 2} x^{2-\alpha_M},$$  \hspace{1cm} (10)

$$\Xi_2 (x) = \frac{2\pi \lambda_P \beta}{\alpha_M - 2} \left( R_M (x) \right)^{2-\alpha_M} \left( P_0 \beta^{-\eta} \int_0^{r_o} t^{\alpha_P - \eta} f_{|X_{o,P}|} (t) dt + P_{\max} \int_{r_o}^\infty f_{|X_{o,P}|} (t) dt \right)$$  \hspace{1cm} (11)

$$F_{|X_{o,P}=x|} (\gamma) = \exp \left\{ -\frac{\delta^2 x^{\alpha_P - \gamma}}{\beta P_{o,P}} - \frac{\lambda_M 2\pi x^{\alpha_P - \gamma} P_{u,M}}{P_o,P} \left( R_p (x) \right)^{2-\alpha_P} \right\} \frac{\alpha_P}{\alpha_P - 2} \left[ e^{\alpha_P - 2/\alpha_P}, 1; 2 - \frac{2}{\alpha_P}; -\frac{x^{\alpha_P - \gamma}}{P_{o,P}} P_{u,M} (R_p (x))^{-\alpha_P} \right]$$

$$- \lambda_P 2\pi^2 P_{o,P} x^{2-\alpha_P} \left( P_{o,P} \right)^{-\eta} \int_0^{r_o} t^{\alpha_P - \eta} F_1 \left[ \left( \frac{\alpha_P - 2}{\alpha_P}, 1; 2 - \frac{2}{\alpha_P}; -\frac{x^{\alpha_P - \gamma}}{P_{o,P}} P_{\max} x^{-\alpha_P} \right) f_{|X_{o,P}|} (t) dt \right]$$

$$- \lambda_P 2\pi^2 P_{o,P} x^{2-\alpha_P} \left( P_{o,P} \right)^{-\eta} \int_0^{r_o} t^{\alpha_P - \eta} F_1 \left[ \left( \frac{\alpha_P - 2}{\alpha_P}, 1; 2 - \frac{2}{\alpha_P}; -\frac{x^{\alpha_P - \gamma}}{P_{o,P}} P_{\max} x^{-\alpha_P} \right) f_{|X_{o,P}|} (t) dt \right]$$  \hspace{1cm} (12)

### IV. NUMERICAL RESULTS

In this section, numerical results are presented to evaluate the area uplink spectrum efficiency of MUE and PUE in the massive MIMO aided HCN. The density of MBSs is $\lambda_M = (500^2 \times \pi)^{-1} \text{m}^{-2}$ in a circular region with radius $1 \times 10^4 \text{m}$. Each MBS is equipped with $N = 200$ antennas. Such a network is assumed to operate at a carrier frequency of 1 GHz, the path loss exponents are $\alpha_M = 3.5$ and $\alpha_P = 4$, the MBS transmit power is $P_M = 43 \text{dBm}$, the PBS transmit power is $P_B = 30 \text{dBm}$, the MUE transmit power is set as 15 dBm, and the maximum PUE transmit power is $P_{\max} = 23 \text{dBm}$. The power density is $P_o = \text{-}80 \text{dBm}$ [13]. In the figures, Monte Carlo (MC) simulated values of the uplink spectrum efficiency marked by ‘o’ are numerically obtained to validate the analysis.

Fig. 1 shows the area uplink spectrum efficiency versus density ratio between PBS and MBS. The number of users served in each macrocell is $S = 20$ and the biasing factor is $B = 3 \text{dB}$. We see that uplink power control applied in the picocells can significantly affect the area uplink spectrum efficiency of the macrocell tier and picocell tier. Specifically, when the pathloss compensation factor is lower, uplink performance in the macrocells is enhanced, due to the fact that macrocells experience less interference from the uplink transmissions in the picocells. Although users in the picocells can use the maximum transmit power $P_{\max} = 23 \text{dBm}$ when they are not close to their serving PBSs, the area uplink spectrum efficiency of the picocell tier is declined. The reason is that the transmit power of the users who are very close to their serving PBSs is controlled, and their achievable uplink spectrum efficiency decreases significantly. Deploying more PBSs improves the area uplink spectrum efficiency of picocell tier, however, the uplink performance of the macrocell tier decreases due to more uplink interference from the picocell tier.

Fig. 2 shows the the area uplink spectrum efficiency versus...
number of users served by each MBS. The density of PBSs is \( \lambda_P = 30 \times 10^{-5} \), and the power control compensation factor \( \eta = 0.8 \). We find that CRE has a substantial effect on the area uplink spectrum efficiency of the macrocell tier, when more users per macrocell are served. However, the effect of CRE on the uplink performance of picocell tier is not significant, since users in the picocells are allowed to use large transmit power (the maximum \( P_{\text{max}} = 25 \text{ dBm} \)) when they are comparatively far from their serving PBSs. Obviously, serving more users increases the area spectrum efficiency of the macrocell tier due to more multiplexing gains, and more uplink transmissions in the macrocells result in a mild decrease in the area uplink spectrum efficiency of the picocell tier, since interference from MUEs increases.

V. CONCLUSION

We considered uplink interference mitigation in the massive MIMO aided heterogeneous cellular networks. By jointly adopting uplink power control and cell range expansion, our results confirmed that uplink inter-tier interference can be effectively coordinated.

APPENDIX A: A PROOF OF THEOREM 1

Based on (4), the area uplink spectrum efficiency is calculated as

\[
C_M = S \lambda_M \mathbb{E} \left( \log_2 \left( 1 + \text{SINR}_M \right) \right). \tag{A.1}
\]

By using Jensen’s inequality, a lower bound for \( \Theta_1 \) is

\[
\Theta_1^L = \log_2 \left( 1 + e^{Z_1 + Z_2} \right), \tag{A.2}
\]

where

\[
\begin{align*}
Z_1 = & \mathbb{E} \left\{ \ln \left( P_{u,M} h_{o,M} \beta |X_{o,M}|^{-\alpha_M} \right) \right\}, \\
Z_2 = & \mathbb{E} \left\{ \ln \left( \frac{1}{I_M + I_P + \delta^2} \right) \right\}. \tag{A.3}
\end{align*}
\]

We first calculate \( Z_1 \) as

\[
Z_1 = \ln \left( P_{u,M} \beta \right) + \mathbb{E} \left\{ \ln \left( h_{o,M} \right) \right\} - \alpha_M \mathbb{E} \left\{ \ln \left| X_{o,M} \right| \right\}. \tag{A.4}
\]

Considering that \( h_{o,M} \sim \Gamma(N - S + 1) \), \( \mathbb{E} \left\{ \ln \left( h_{o,M} \right) \right\} \) is calculated as \( \mathbb{E} \left\{ \ln \left( h_{o,M} \right) \right\} = \psi(N - S + 1) \). We next derive \( \mathbb{E} \left\{ \ln \left| X_{o,M} \right| \right\} \) as

\[
\mathbb{E} \left\{ \ln \left| X_{o,M} \right| \right\} = \int_0^{\infty} \ln (x) f_{|X_{o,M}|}(x) \, dx \tag{A.5}
\]

By using Jensen’s inequality, we derive the lower bound on the \( Z_2 \) as

\[
Z_2 > Z_2^L = \ln \left( \frac{1}{\mathbb{E} \{ I_M \} + \mathbb{E} \{ I_P \} + \delta^2} \right). \tag{A.6}
\]

Then, we have

\[
\mathbb{E} \{ I_M \} = \int_0^{\infty} \mathbb{E} \{ I_M \} \left| X_{o,M} = x \right. f_{|X_{o,M}|}(x) \, dx \tag{A.7}
\]

where (a) results from using the Campbell’s theorem [14]. Similarly, \( \mathbb{E} \{ I_P \} \) can also be obtained. Substituting (A.3) and (A.2) into (A.1), we get (13).

REFERENCES